

# **Artificial Intelligence and Machine Learning**

## **Neural Networks**

# Lecture Outline

- Basics Steps in Machine Learning
- Neural Networks
  - Why Neural Networks?
  - Forward pass
  - Backward pass

# Basics Steps in Machine Learning



Data

Model/Algorithm

Loss Function

Optimization Algorithm



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# Introduction to Deep Learning

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## What is a Neural Network?

# Why Neural Networks (Motivation)



Linear decision boundary

# Why Neural Networks (Motivation)



Feature Engineering > Non-linear decision boundary

$X_1$       |       $X_2$

# Why Neural Networks (Motivation)



We need to find the effective  
non-linear combinations



Neural Networks



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# Introduction to Deep Learning

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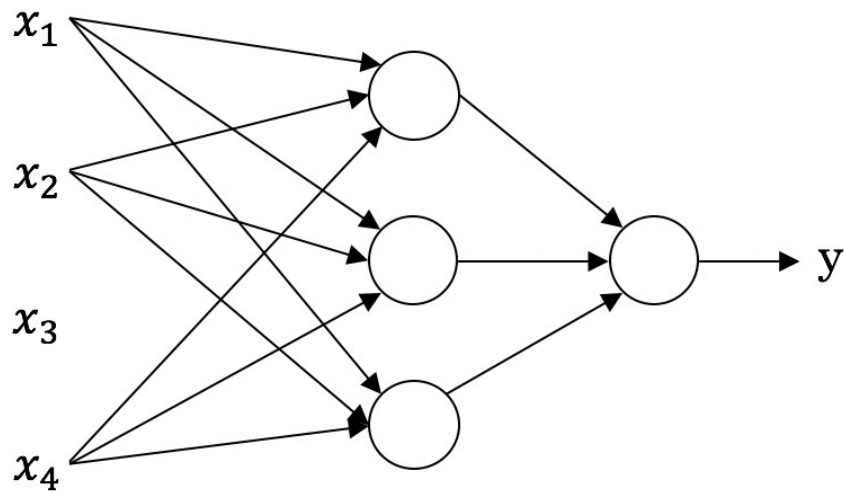
## Supervised Learning with Neural Networks



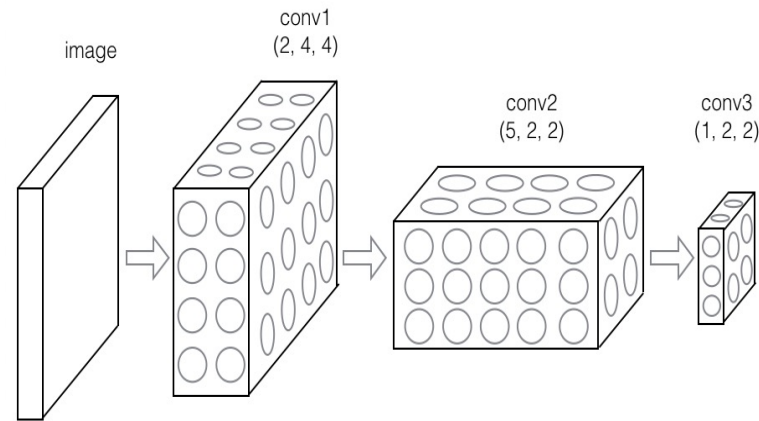
# Supervised Learning

Input(x)	Output (y)	Application
Home features	Price	Real Estate
Ad, user info	Click on ad? (0/1)	Online Advertising
Image	Object (1,...,1000)	Photo tagging
Audio	Text transcript	Speech recognition
English	Chinese	Machine translation
Image, Radar info	Position of other cars	Autonomous driving

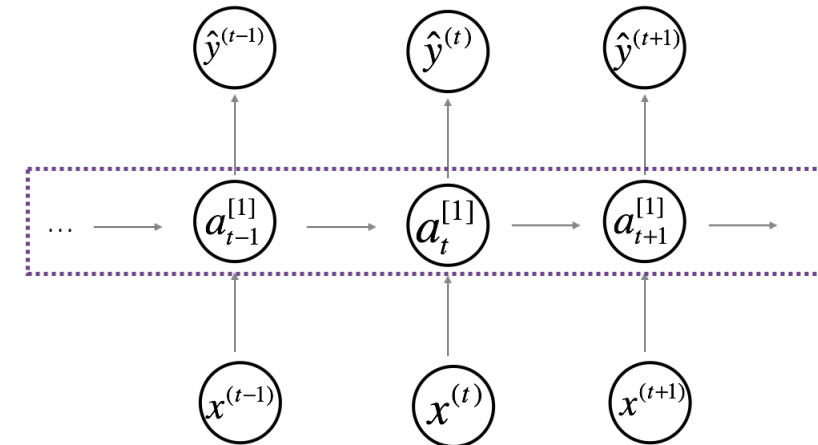
# Neural Network examples



Standard NN



Convolutional NN



Recurrent NN



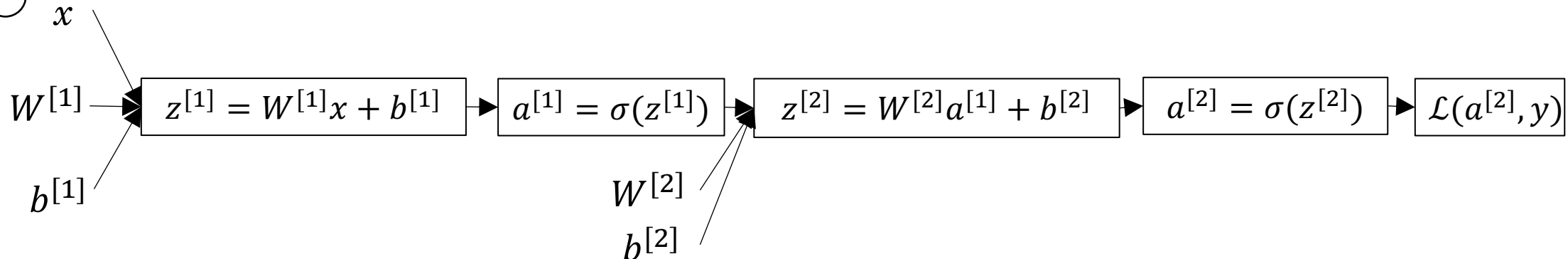
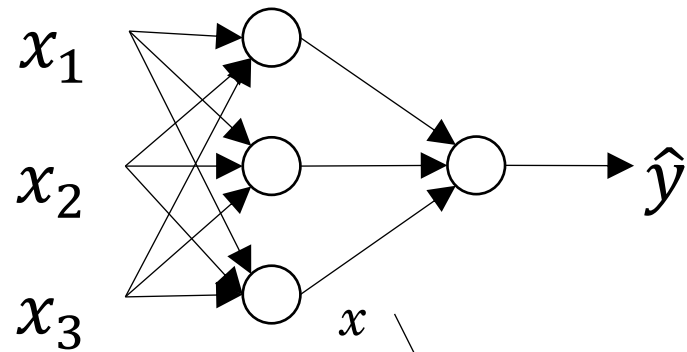
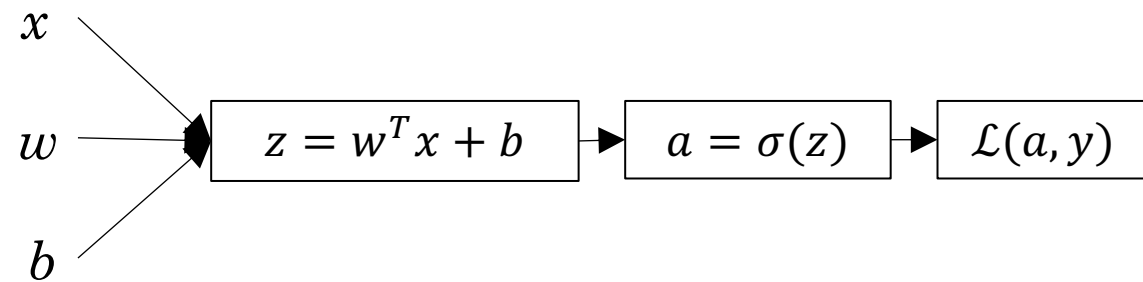
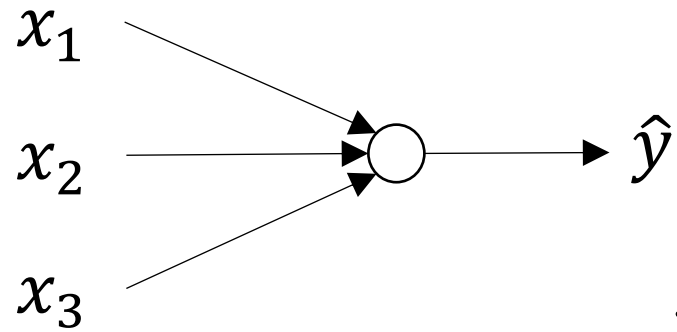
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# One hidden layer Neural Network

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## Neural Networks Overview

# What is a Neural Network?





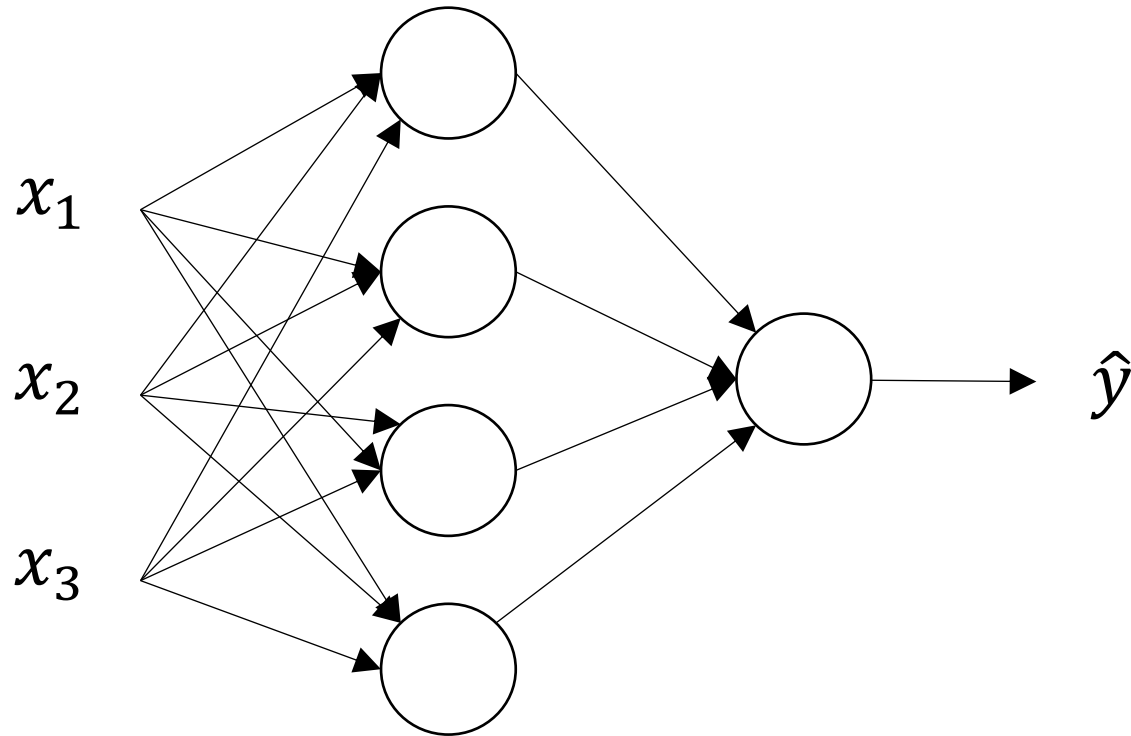
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# One hidden layer Neural Network

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## Neural Networks

# Neural Network Representation



# Neural Networks: Data



# Neural Networks: Model





# Neural Networks: Loss

# Neural Networks: Optimization





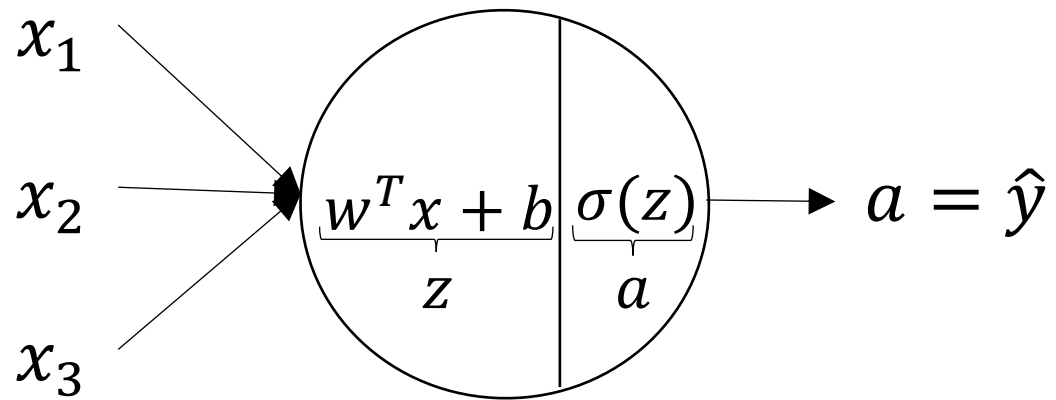
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# One hidden layer Neural Network

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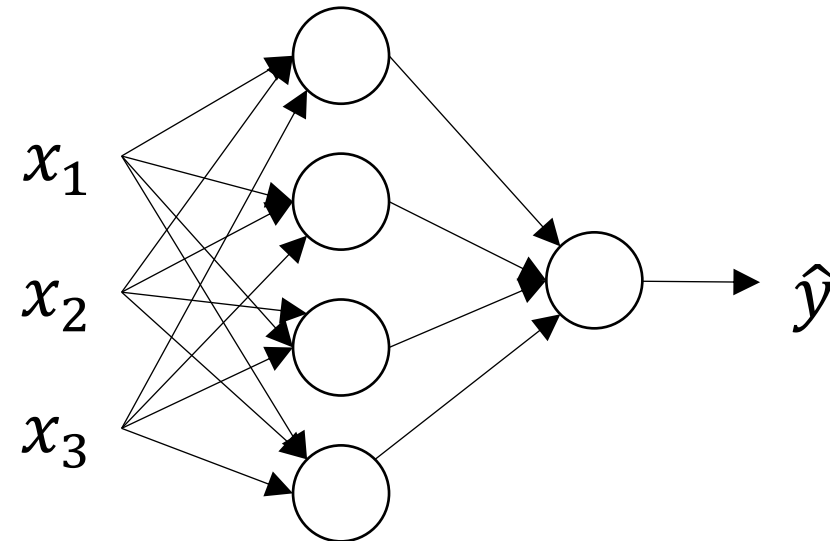
## Computing a Neural Network's Output

# Neural Network Representation

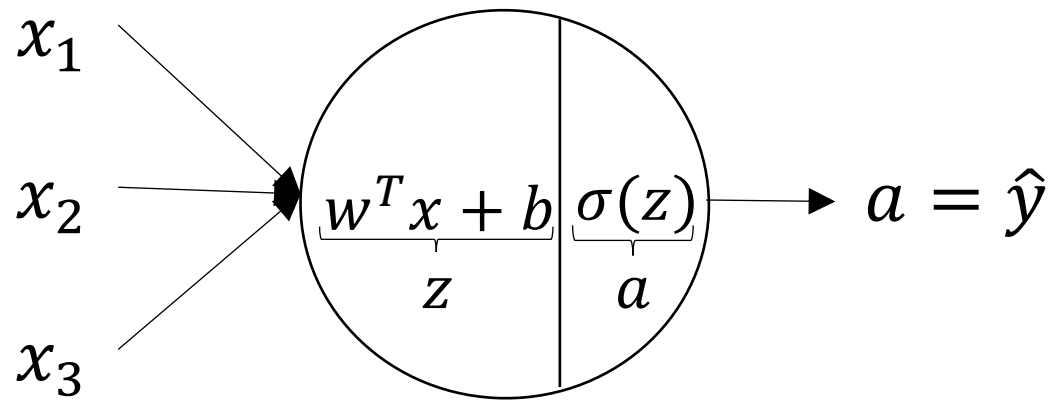


$$z = w^T x + b$$

$$a = \sigma(z)$$

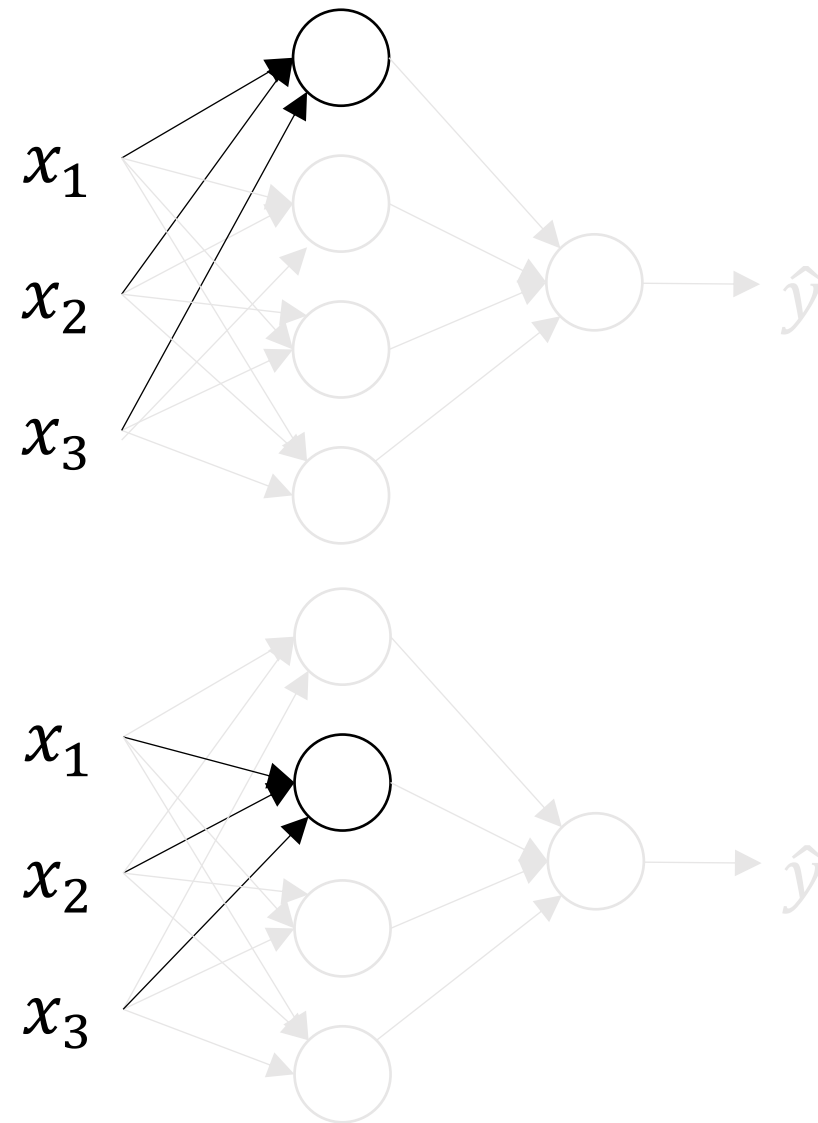


# Neural Network Representation

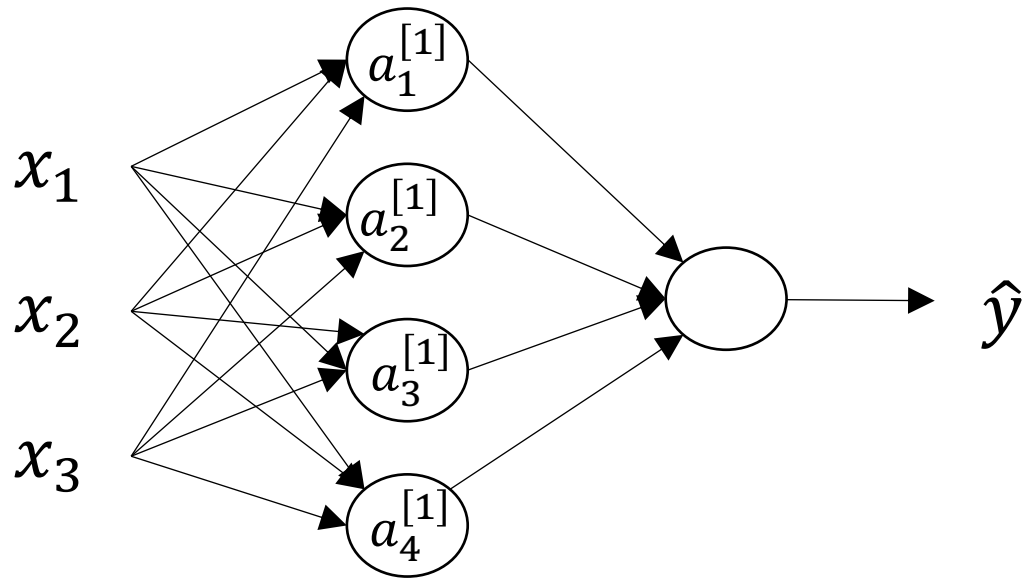


$$z = w^T x + b$$

$$a = \sigma(z)$$



# Neural Network Representation



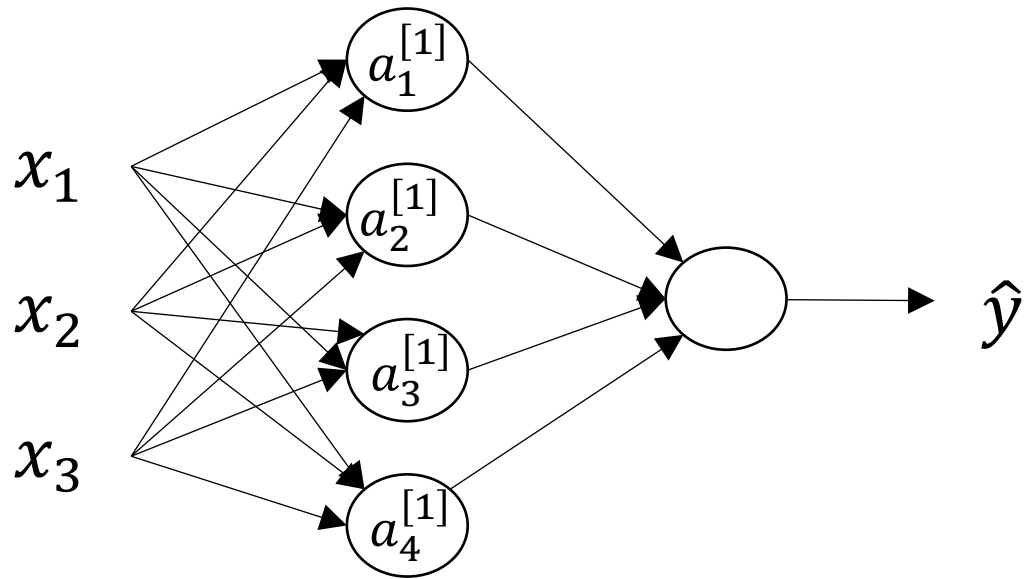
$$z_1^{[1]} = w_1^{[1]T} x + b_1^{[1]}, \quad a_1^{[1]} = \sigma(z_1^{[1]})$$

$$z_2^{[1]} = w_2^{[1]T} x + b_2^{[1]}, \quad a_2^{[1]} = \sigma(z_2^{[1]})$$

$$z_3^{[1]} = w_3^{[1]T} x + b_3^{[1]}, \quad a_3^{[1]} = \sigma(z_3^{[1]})$$

$$z_4^{[1]} = w_4^{[1]T} x + b_4^{[1]}, \quad a_4^{[1]} = \sigma(z_4^{[1]})$$

# Neural Network Representation learning



Given input  $x$ :

$$z^{[1]} = W^{[1]}x + b^{[1]}$$

$$a^{[1]} = \sigma(z^{[1]})$$

$$z^{[2]} = W^{[2]}a^{[1]} + b^{[2]}$$

$$a^{[2]} = \sigma(z^{[2]})$$



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# One hidden layer Neural Network

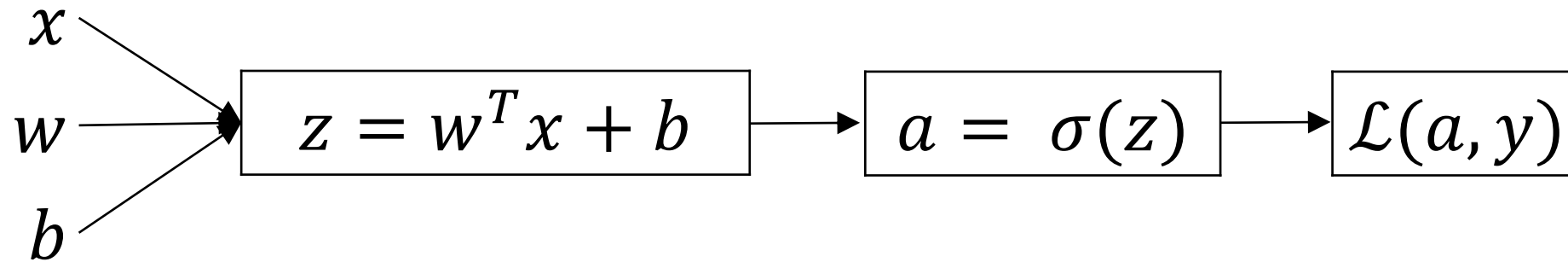
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## Backpropagation intuition

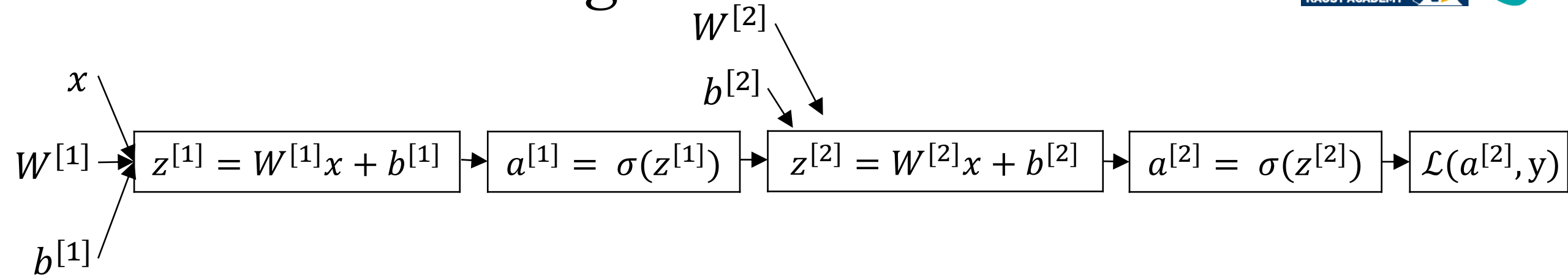


# Computing gradients

## Logistic regression



# Neural network gradients



# Summary of gradient descent

$$dz^{[2]} = a^{[2]} - y$$

$$dW^{[2]} = dz^{[2]} a^{[1]T}$$

$$db^{[2]} = dz^{[2]}$$

$$dz^{[1]} = W^{[2]T} dz^{[2]} * g^{[1]'}(z^{[1]})$$

$$dW^{[1]} = dz^{[1]} x^T$$

$$db^{[1]} = dz^{[1]}$$

# Summary of gradient descent



$$dz^{[2]} = a^{[2]} - y$$

$$dW^{[2]} = dz^{[2]} a^{[1]T}$$

$$db^{[2]} = dz^{[2]}$$

$$dz^{[1]} = W^{[2]T} dz^{[2]} * g^{[1]'}(z^{[1]})$$

$$dW^{[1]} = dz^{[1]} x^T$$

$$db^{[1]} = dz^{[1]}$$

$$dZ^{[2]} = A^{[2]} - Y$$

$$dW^{[2]} = \frac{1}{m} dZ^{[2]} A^{[1]T}$$

$$db^{[2]} = \frac{1}{m} np.sum(dZ^{[2]}, axis = 1, keepdims = True)$$

$$dZ^{[1]} = W^{[2]T} dZ^{[2]} * g^{[1]'}(Z^{[1]})$$

$$dW^{[1]} = \frac{1}{m} dZ^{[1]} X^T$$

$$db^{[1]} = \frac{1}{m} np.sum(dZ^{[1]}, axis = 1, keepdims = True)$$



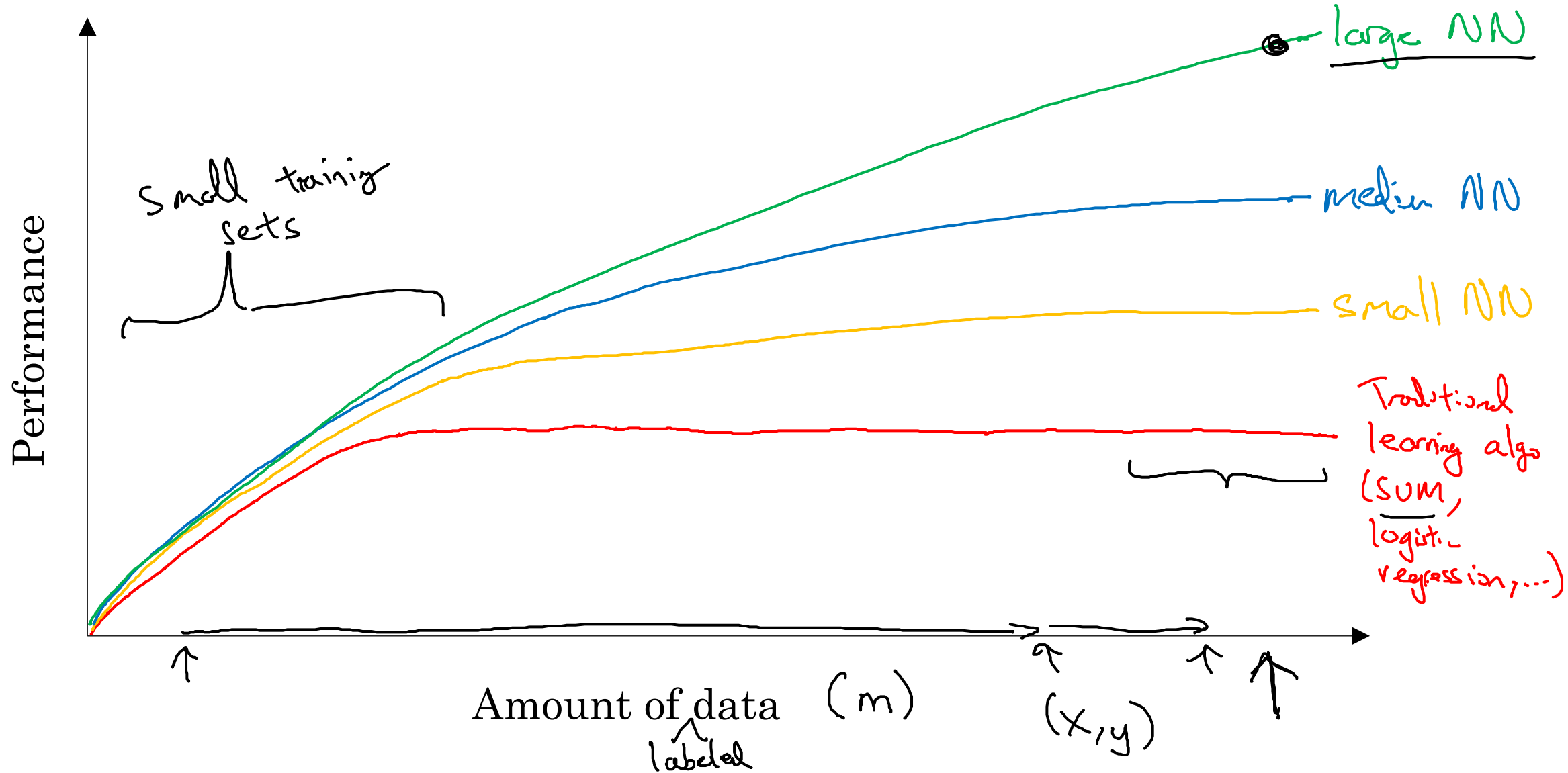
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# Introduction to Neural Networks

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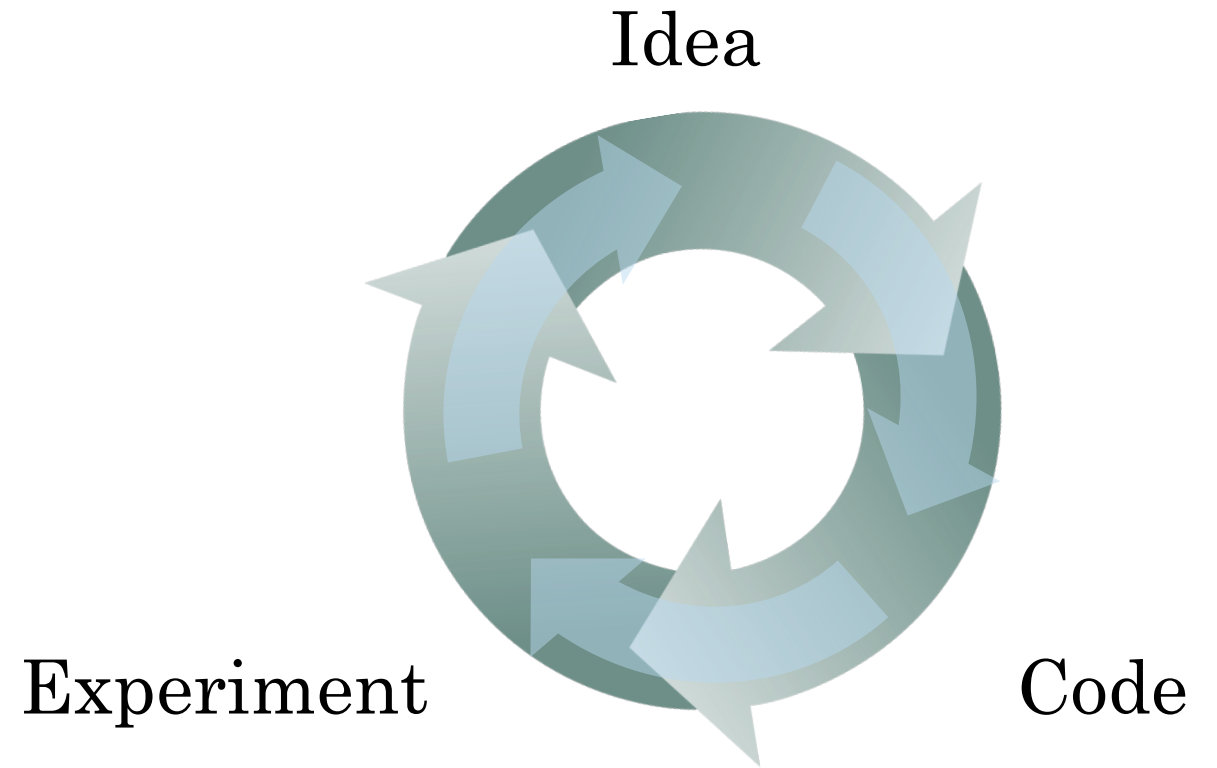
## Why is Deep Learning taking off?

# Scale drives deep learning progress



# Scale drives deep learning progress

- Data
- Computation
- Algorithms





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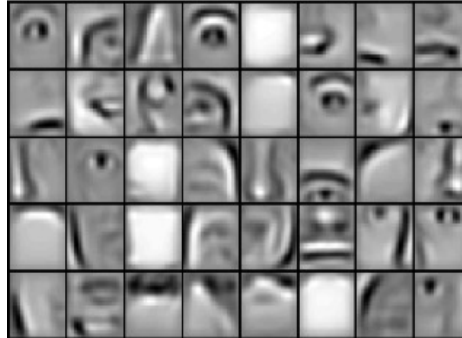
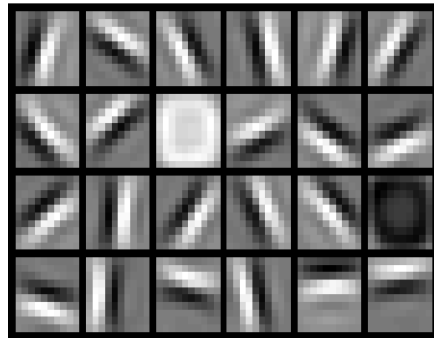
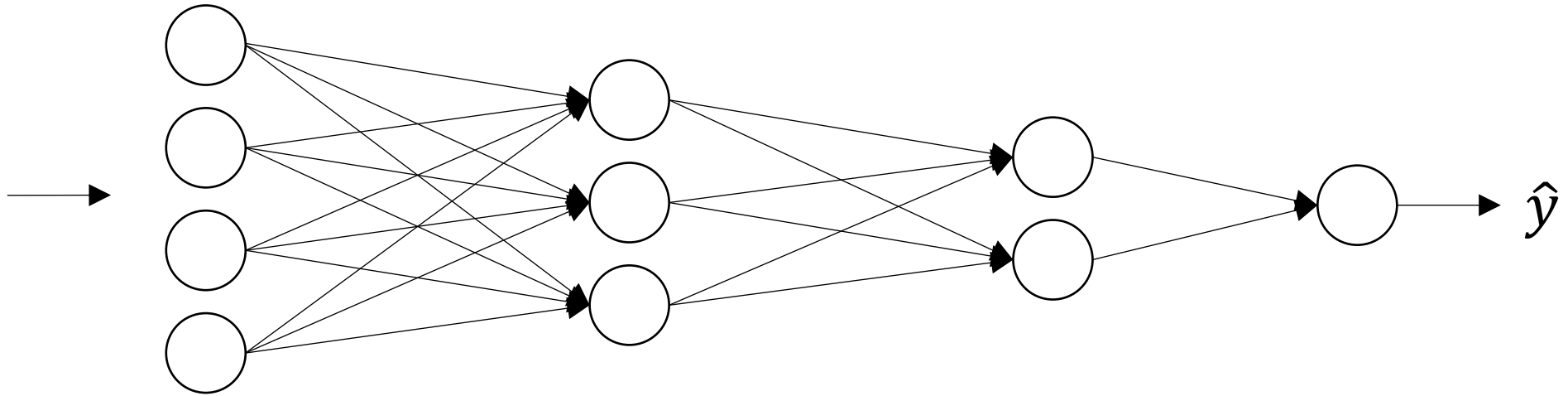
# Deep Neural Networks

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## Why deep representations?



# Intuition about deep representation





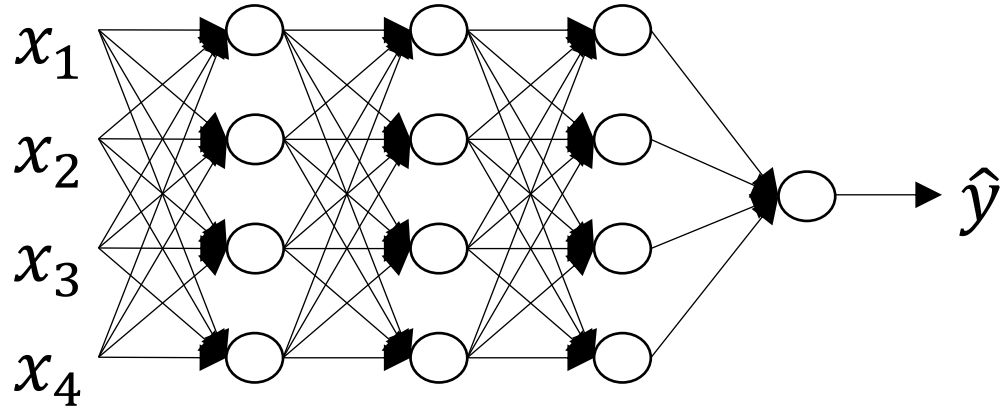
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# Deep Neural Networks

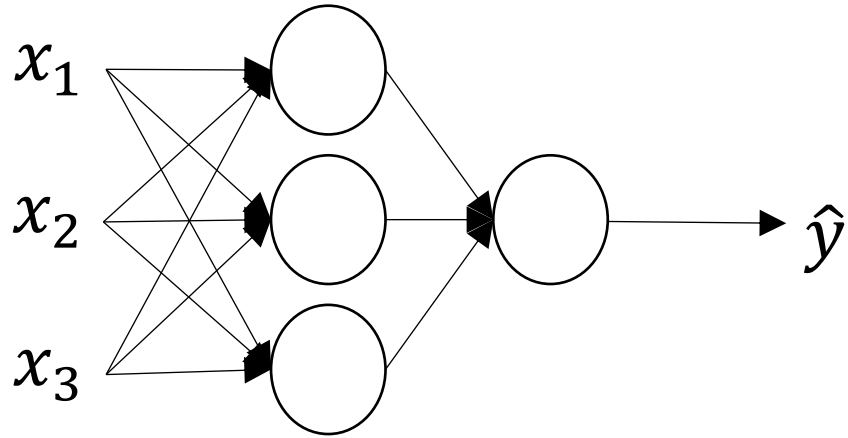
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Building blocks of  
deep neural networks

# Forward and backward functions



# Activation functions



Given  $x$ :

$$z^{[1]} = W^{[1]}x + b^{[1]}$$

$$a^{[1]} = \sigma(z^{[1]})$$

$$z^{[2]} = W^{[2]}a^{[1]} + b^{[2]}$$

$$a^{[2]} = \sigma(z^{[2]})$$

# Pros and cons of activation functions

